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INTERNATIONAL APPLICATION PUBLISHED UNDER THE PATENT COOPERATION TREATY (PCT)

(51) International Patent Classification 6 : G01N 27/27	A1	(11) International Publication Number: WO 99/61902 (43) International Publication Date: 2 December 1999 (02.12.99)
<p>(21) International Application Number: PCT/US99/11636</p> <p>(22) International Filing Date: 26 May 1999 (26.05.99)</p> <p>(30) Priority Data: 60/086,876 27 May 1998 (27.05.98) US 60/088,804 9 June 1998 (09.06.98) US</p> <p>(63) Related by Continuation (CON) or Continuation-in-Part (CIP) to Earlier Applications US 60/086,876 (CON) Filed on 27 May 1998 (27.05.98) US 60/088,804 (CON) Filed on 9 June 1998 (09.06.98)</p> <p>(71) Applicant (for all designated States except US): CALIFORNIA INSTITUTE OF TECHNOLOGY [US/US]; 1201 E. California Boulevard, Pasadena, CA 91125 (US).</p> <p>(72) Inventors; and (75) Inventors/Applicants (for US only): DOLEMAN, Brett, J. [CA/US]; Apartment 106, Four Bayside Village Place, San Francisco, CA 94107 (US). SEVERIN, Erik, J. [US/US]; 1440 Orlando Road, San Marino, CA 91108 (US). LEWIS,</p>	<p>Nathan, S. [US/US]; 5259 Gould Avenue, La Canada, CA 91011 (US).</p> <p>(74) Agents: CHAN, Melvin, D. et al.; Townsend and Townsend and Crew LLP, 8th floor, Two Embarcadero Center, San Francisco, CA 94111 (US).</p> <p>(81) Designated States: AE, AL, AM, AT, AU, AZ, BA, BB, BG, BR, BY, CA, CH, CN, CU, CZ, DE, DK, EE, ES, FI, GB, GD, GE, GH, GM, HR, HU, ID, IL, IN, IS, JP, KE, KG, KP, KR, KZ, LC, LK, LR, LS, LT, LU, LV, MD, MG, MK, MN, MW, MX, NO, NZ, PL, PT, RO, RU, SD, SE, SG, SI, SK, SL, TJ, TM, TR, TT, UA, UG, US, UZ, VN, YU, ZA, ZW, ARIPO patent (GH, GM, KE, LS, MW, SD, SL, SZ, UG, ZW), Eurasian patent (AM, AZ, BY, KG, KZ, MD, RU, TJ, TM), European patent (AT, BE, CH, CY, DE, DK, ES, FI, FR, GB, GR, IE, IT, LU, MC, NL, PT, SE), OAPI patent (BF, BJ, CF, CG, CI, CM, GA, GN, GW, ML, MR, NE, SN, TD, TG).</p> <p>Published <i>With international search report.</i></p>	
<p>(54) Title: METHOD OF RESOLVING ANALYTES IN A FLUID</p> <p>(57) Abstract</p> <p>A statistical metric, based on the magnitude and standard deviations along linear projections of clustered array response data, is utilized to facilitate an evaluation of the performance of detector arrays in various vapor classification tasks. This approach allows quantification of the ability of arrays of different types including carbon black-insulating polymer composite chemiresistor sensors, tin oxide sensors and bulk conducting organic polymer sensors to distinguish between analytes. The evaluation of questions such as the optimal number of detectors required for a specific task, whether improved performance is obtained by increasing the number of detectors in a detector array, and how to assess statistically the diversity of a collection of detectors in order to understand more fully which properties are underrepresented in a particular set of array elements, are addressed.</p>		

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Method Of Resolving Analytes In a Fluid

STATEMENT AS TO RIGHTS TO INVENTIONS MADE UNDER FEDERALLY SPONSORED RESEARCH AND DEVELOPMENT

5 The United States Government may have certain rights to the present invention pursuant to Grant No. DAAG55-97-1-0187 awarded by the U.S. Army and Grant No. DAAK60-97-K-9503 awarded by the U.S. Navy.

RELATED CASES

10 The present patent application claims priority from U.S. Provisional Application No. 60/086,876, filed May 27, 1998 and from U.S. Provisional Application No. 60/088,804, filed June 9, 1998.

BACKGROUND OF THE INVENTION

15 The present invention is related to the electrochemical sensing of analytes in a medium or a vacuum, such as odors, smells, vapors and the like in air, and in particular, to methods of determining or resolving such analytes once detected.

 There has been considerable interest in developing devices for the general detection of analytes in a fluid, vacuum, air, or other medium. A specific instance of an
20 analyte detector is a device for sensing smells or odors (i.e., analytes in air). It is well recognized that some animals, such as dogs, have a much keener sense of smell than human beings. Because of the acuity of their noses, dogs have been utilized for many tasks including, for example, the detection of bombs, mines, drugs, poison gases, and illegal contraband; dogs also aid in the search and rescue of humans. Devices for sensing
25 smell would be useful for the applications where animals are traditionally used, as well as for a multitude of uses where animals are impractical or inappropriate.

 Moreover, devices for the general detection of analytes have potentially many more applications than devices for detecting only smells and odors. For example, the uses for devices for analyte detection include the detection of chemical leaks, quality
30 control in food processing, medical diagnosis and testing, fabrication and manufacture of commercial and industrial goods, pharmaceutical production, testing or evaluating any odorant or analyte in any medium (e.g., fuel, oil, wine, solvents), and many other applications. An instrument for analyte detection would be highly desirable in many

industries and applications, such as the chemical, petrochemical, food, fragrance, medical, automotive, military, environmental, health and safety, and air quality monitoring sectors. Therefore, it is desirable to develop techniques and devices for the detection of analytes.

One approach for sensing analytes is the use of an array of sensors on a substrate. Each sensor has a chemical compound which has a particular electrical characteristic response to exposure to an analyte, such as an odor. The sensors are connected to electrical circuits by which the responses of the chemical compounds to the analyte are retrieved. See, for example, U.S. Appln. No. 09/130,775, entitled "SENSOR ARRAYS ON AN IC FOR ANALYTE DETECTION," filed August 7, 1998 and assigned to the present assignee, and which is incorporated by reference herein. Each site responds to a broad class of stimuli, with the collective response of the many different members of the sensor array providing a fingerprint of an analyte of interest. The number of data points generated for each analyte is large. Even with modern high-speed computers, the amount of data generated may significantly slow operations to analyze the sensed analytes under real world constraints, in particular. Therefore, there is a need for techniques and systems which are computationally efficient in determining or resolving an analyte upon detection from the large amount of data derived from the sensor array.

The present invention is directed toward such a technique and system for quickly and efficiently resolving analytes. A quantitative metric is generated for the effective resolution of different analytes by the sensor array. The present invention further provides a powerful tool in the analysis and construction of olfactory sensor arrays.

SUMMARY OF THE INVENTION

The present invention provides for a method for distinguishing different odors by the steps of disposing a plurality of d sensors in an array, each sensor having different electrical responses to different odors; exposing the sensors to first and second odors; generating first and second sets of data points from each of the sensors, each set corresponding to the first or second odor, each data point being represented by a vector in a d -dimensional space; determining an axis in the d -dimensional space, the axis having the property that projections of the data points onto the axis in the d -dimensional space have optimal separation; and resolving the first odor from the second odor by the separation.

BRIEF DESCRIPTION OF THE DRAWINGS

Fig. 1 illustrates an array of sensors connected to a computer unit, in accordance with one aspect of the present invention;

Fig. 2 illustrates a typical electrical response of a sensor to the presence of an analyte;

Fig. 3 is a two-dimensional representation of the resolution factor obtained by the projection of the data points of two analytes upon optimal projection vector, according to one aspect of the present invention;

Fig. 4A is a table of pairwise resolution factors obtained from an array of fourteen carbon black polymer-composite detectors for nineteen solvent analytes; Fig. 4B is a table of pairwise resolution factors obtained from an array of eight tin oxide detectors for the nineteen solvent analytes; Fig 4C is a table of pairwise resolution factors obtained from an array of twelve bulk conducting organic polymer detectors for the nineteen solvent analytes;

Fig. 5 is a table of normalized pairwise resolution factors obtained from the array of fourteen carbon black-polymer composite detectors;

Fig. 6 is a table of principal components of the fourteen element carbon black-polymer composite detector response data;

Fig. 7A is a plot of the average ability to resolve all analyte pairs, $\overline{rf}(n)$, versus the number n of sensors in the array; Fig. 7B is a plot of the average ability to resolve all analyte pairs, $\overline{rf}(n)$, for the carbon black-polymer composite detectors versus the number n of sensors in the array using normalized data; Fig. 7C is a plot of the maximum average resolution factor, $\overline{rf}_{\max}(n)$, the specific set of specified number n of detectors which have the largest ability on average to resolve all solvent vapor analyte pairs versus the number of detectors n ; Fig. 7D is a plot of the maximum average resolution factor, $\overline{rf}_{\max}(n)$, for the carbon black-polymer composite detectors versus the number of detectors n using normalized data;

Fig. 8A is a plot of the smallest pairwise resolution factor, rf_{worst} , versus the number n of sensors in the array; Fig. 8B is a plot of the smallest pairwise resolution factor, rf_{worst} , versus the number n of carbon black-polymer composite sensors in the array;

Fig. 9 is a plot of the largest solution factor, $rf_{\text{best}}(n)$, to resolve ethyl acetate from tetrahydrofuran at fixed concentrations for all combinations of detectors

from the number n of the detectors in the array versus the number n of the detectors in the array; and

Fig. 10 is a table of the results of resolving on average and resolving the worst-resolved analyte pair using combined arrays of carbon black-polymer composite, tin oxide, and conducting organic polymer sensors.

DESCRIPTION OF THE SPECIFIC EMBODIMENTS

In accordance with the present invention, an array 10 of sensors is disposed on a substrate, such as shown in Fig. 1. Each sensor is responsive to different odor analytes in a particular manner so that the response of all the sensors creates a "fingerprint" which can be used to identify the analyte. These sensors may include surface acoustic wave (SAW) devices, tin oxide detectors, conducting organic polymers, dye-impregnated polymer films on fiber optic detectors, polymer-coated micromirrors, quartz crystal microbalances (QCMs), electrochemical gas detectors, chemically sensitive field-effect transistors, carbon black-polymer composite chemiresistors, micro-electro-mechanical system (MEMs) devices, and micro-opto-electro-mechanical system (MOEMs) devices. If the devices are compatible, the sensors might be placed on a single substrate, such as that of semiconductor material, for the benefits of miniaturization and convenience. Each of the sensors is connected by wires through an array interface 12 to a computer unit 1 which receives the electrical responses from the sensors upon exposure to an odor analyte. The computer unit is programmed to save and process the data in accordance with the data processing steps set forth below. The results may be displayed on a screen 5 of the computer unit's monitor and/or saved in memory, such as represented by a computer disk 15.

With d sensors 11 in the array 10, each sampling of the array creates d values, i.e., each sampling data point can be considered a d -dimensional vector. If there are eight sensors, each data point has eight values, one from each sensor. Hence it can be readily understood that the amount of data generated by the array 10 can be quite large. For example, if there are 100 sensors in the array 10, then each data point has 100 associated values. Furthermore, there may be several samples taken for each analyte thereby magnifying the computational problem of analysis. The present invention reduces the problem of d -dimensional data point vectors to the comparison of values in one dimension. This is performed by the application of the Fisher linear discriminant to

the data. In this manner the resolution of different odor analytes can be made efficiently and effectively with a quantitative metric.

In the description below, specific types of sensors or detectors (tin oxide, conducting organic polymer and carbon black-polymer composite detectors) are described with respect to the detection of specific analytes (some nineteen different solvent vapors including 1,2-dimethoxyethane, acetone, acetonitrile, anisole, benzene, butylamine, chloroform, cyclohexane, dichloromethane, ethanol, ethyl acetate, isopropanol, methanol, n-heptane, n-pentane, tetrahydrofuran, toluene, triethylamine and α,α,α -trifluorotoluene). It should be understood that the present invention described herein is applicable to other sensors, including those of different sensing modalities, and analytes.

The present invention might best be understood by assuming a typical experiment where it is desired to obtain a quantitative measure of the efficiency of an array of 14 detectors in distinguishing between two solvents. The array is exposed to each solvent for 12 cycles. Each cycle consists of 300 seconds of pure carrier gas over the detector array, then 240 seconds of gas and the solvent, then 300 seconds of pure carrier gas again. This procedure ensures steady state conditions before, during, and after, exposure to the solvent mixture. The resistance of each detector, j , is measured at 3 second intervals throughout each exposure, i . The experiment data base for this typical experiment thus consists of $2 \times 12 \times 840 / 3 \times 14 = 94,080$ resistance measurements.

The first step in processing these data finds the maximum relative resistance change during each exposure for each solvent and each detector. Of course, it should be understood that the present invention is also applicable to other electrical parameters, such as voltage, current, capacitance, or inductance, which might vary in response to the exposure to analytes. A typical detector response during an exposure will have the general form, such as illustrated in Fig. 2. The baseline resistance, R_b , is the stable value attained by the detector just prior to the introduction of the solvent into the carrier gas stream, that is, the 100th measurement (300 seconds sampled at 3 second intervals). The maximum resistance value attained during the gas and solvent interval, R_{max} , is readily obtained between measurements 101 and 540 (typically towards the end of the 240 second gas and solvent interval). The maximum relative resistance change for the cycle is then:

$$(R_{max} - R_b)/R_b.$$

The experimental database has now been reduced to $2 \times 2 \times 14 = 336$ measurements. These can be regarded as a set of twelve 14-dimensional points, α_{ij} ($i = 1$ to 12 and $j = 1$ to 14), for each solvent. That is:

$$\alpha_{ij} = \begin{pmatrix} \alpha_{i1} \\ \alpha_{i2} \\ \vdots \\ \alpha_{i,14} \end{pmatrix}, \begin{pmatrix} \alpha_{21} \\ \alpha_{22} \\ \vdots \\ \alpha_{2,14} \end{pmatrix}, \dots, \begin{pmatrix} \alpha_{12,1} \\ \alpha_{12,2} \\ \vdots \\ \alpha_{12,14} \end{pmatrix}$$

5

A quantitative measure of the array efficiency in distinguishing between the solvents is obtained using the optimum resolution factor (defined by Fisher's linear discriminant):

$$10 \quad rf = \frac{d_{\bar{w}}}{\sqrt{\sigma_{a,\bar{w}}^2 + \sigma_{b,\bar{w}}^2}}$$

The resolution factor, rf, is from an optimal projection vector, \bar{w} , where

$$\bar{w} = S_w^{-1}(\bar{m}_a - \bar{m}_b)$$

15 where \bar{m}_a and \bar{m}_b are the respective 14-dimensional means of each of data sets of the two solvents, that is:

$$\bar{m} = \begin{pmatrix} m_1 \\ m_2 \\ \vdots \\ m_{14} \end{pmatrix} = \frac{1}{12} \begin{pmatrix} \sum \alpha_{i1} \\ \sum \alpha_{i2} \\ \vdots \\ \sum \alpha_{i,14} \end{pmatrix} = \frac{1}{12} \begin{pmatrix} \alpha_{11} + \alpha_{21} + \dots + \alpha_{12,1} \\ \alpha_{12} + \alpha_{22} + \dots + \alpha_{12,2} \\ \vdots \\ \alpha_{1,14} + \alpha_{2,14} + \dots + \alpha_{12,14} \end{pmatrix}$$

20

The within-class scatter matrix, S_w , is the sum of the matrices S_a and S_b where

$$S = \sum (x_{ij} - \mathbf{m})(x_{ij} - \mathbf{m})^T$$

Each S is the sum over the 12 exposures of 14×14 - dimensional terms, such as:

5

$$\begin{pmatrix} \alpha_{11} - m_1 \\ \alpha_{12} - m_2 \\ \cdot \\ \cdot \\ \cdot \\ \alpha_{1,14} - m_{14} \end{pmatrix} (\alpha_{11} - m_1 \quad \alpha_{12} - m_2 \quad \cdots \quad \alpha_{1,14} - m_{14})$$

10 Then

$$S_w = S_a + S_b ; \text{ and}$$

$$\mathbf{w} = S_w^{-1} (\mathbf{m}_a - \mathbf{m}_b)$$

d_w is the distance along the vector \mathbf{w} between the projected means and is computed as:

$$d_w = [\mathbf{w} * (\mathbf{m}_a - \mathbf{m}_b)(\mathbf{m}_a - \mathbf{m}_b)^T]^{1/2}$$

15

Also, the denominator for rf is:

$$\mathbf{w}^T S_w \mathbf{w} = \sqrt{\sigma_{a,w}^2 + \sigma_{b,w}^2}$$

to obtain the resolution factor, rf , between two solvents.

20

Stated differently, the Fisher linear discriminant searches for the projection vector, \mathbf{w} , in the detector space which maximizes the pairwise resolution factor, i.e., rf , for each set of analytes, and reports the value of rf along this optimal linear discriminant vector. Fig. 3 is a graphical two-dimensional representation of the data points of analytes a and b , and their projected means along the optimal discriminant vector \mathbf{w} . The rf value is an inherent property of the data set and does not depend on whether principal component space or original detector space is used to analyze the response data. This resolution factor is basically a multi-dimensional analogue to the separation factors used

25

to quantify the resolving power of a column in gas chromatography, and thus the rf value serves as a quantitative indication of how distinct two patterns are from each other, considering both the signals and the distribution of responses upon exposure to the analytes that comprise the solvent pair of concern. For example, assuming a Gaussian
5 distribution relative to the mean value of the data points that are obtained from the responses of the array to any given analyte, the probabilities of correctly identifying an analyte as a or b from a single presentation when a and b are separated with resolution factors of 1.0, 2.0 or 3.0 are approximately 76%, 92% and 98% respectively.

The efficacy of the technique is demonstrated in the empirical data
10 obtained with arrays having different types of detectors, i.e., tin oxide, conducting organic polymer and carbon black-polymer composite detectors. In the experiments, the array of tin oxide detectors had eight different detectors; the array of conducting organic polymer detectors had twelve different detectors; and the array of carbon block-polymer composite detectors had fourteen different detectors. Nineteen different solvent vapors (1,2-
15 dimethoxyethane, acetone, acetonitrile, anisole, benzene, butylamine, chloroform, cyclohexane, dichloromethane, ethanol, ethyl acetate, isopropanol, methanol, n-heptane, n-pentane, tetrahydrofuran, toluene, triethylamine and α,α,α -trifluorotoluene) were used as the analytes.

Figs. 4A-4C display the pairwise resolution factors, rf , for the three
20 different types of detector arrays. The table in Fig. 4A presents the resolution factors obtained from an array of the fourteen carbon black-polymer composite detectors for all 171 pairs of the nineteen vapors. The tables in Figs. 4B and 4C display similar data for the eight tin oxide detectors and for the twelve available bulk conducting organic polymer detectors which had all been exposed to the nineteen test solvent vapors.

25 These resolution factors allow assessment of the performance of the different arrays, or of the performance of subsets of the arrays, in various sensing tasks. Two criteria are chosen as measures of array performance: (1) the average ability to resolve all analyte pairs, \overline{rf} , and (2) the ability of the array to resolve the worst-resolved analyte pair, rf_{worst} . The \overline{rf} 's and standard deviations across all analyte pairs for the full
30 carbon black-polymer composite, tin oxide, and conducting polymer arrays (Figs. 4A-4C) evaluated were 145 ± 93 , 27 ± 23 and 18 ± 16 respectively, while the rf_{worst} values were 23, 4.0 and 1.4, respectively. Based on this analysis, the carbon black-polymer composite array yielded the largest mean statistical separation of the response patterns produced by

exposure of this particular collection of available detectors to this particular collection of test solvent vapors exposed at fixed concentrations. Using the same experimental conditions, the carbon black composite detector array also yielded the largest separation between the worst-resolved pair of analytes.

5 The present invention also provides a means of determining the resolving power of the carbon black-polymer composite detectors at variable analyte concentrations. The analysis described above serves as one quantitative measure of the classification ability of different detector arrays. However, this metric also includes any separation between data clusters that arises from the differences in the amplitudes of the
10 array response to the vapors in the test analyte set. Two solvents that produce nearly identical patterns, but with very different absolute detector response amplitudes, would produce two well-separated clusters. The large pairwise rf value for such a solvent pair produced by the data analysis method described above is relevant for tasks in which the concentration of an analyte is known, or when only detection of changes over time in an
15 otherwise relatively constant analyte composition is of interest. However, such a data analysis procedure is clearly not appropriate for assessing array classification performance in applications where the concentration of the analyte is not known, or not known to be fixed, in advance.

 A measure of array performance under such conditions requires that the
20 fundamental differences in patterns produced by the various vapors be determined without considering differences in the mean magnitude of the detector responses. Carbon black-polymer composite detectors containing $\geq 20\%$ by weight carbon black have been shown to respond linearly to the concentration of several test analytes over at least an order of magnitude of analyte concentration. Hence, a measure of the concentration-
25 independent ability of the array to classify the various solvents in the test can be obtained for the carbon black-polymer composite detectors by normalizing the array responses. This procedure is equivalent to adding an extra degree of freedom, analyte concentration, to the classification task.

 The normalized response, N_{ij} , of the j^{th} detector in an n detector array to
30 the i^{th} of k exposures of a specific analyte is defined as:

$$N_{ij} = \frac{\Delta R_{ij, \max} / R_b}{\frac{1}{n \cdot k} \sum_{j,i} (\Delta R_{ij, \max} / R_b)},$$

where the summation is over the response of all n detectors to all k exposures of the specific analyte. For the carbon black-polymer composite vapor detectors, a comparison of the 171 pairwise resolution factors obtained from normalized responses (the table in Fig. 5) to those obtained only from raw response data (the table in Fig. 4A) shows that allowing the vapor concentration to be a floating variable slightly reduces, but does not remove, the array's ability to resolve the various test analytes at the concentrations used in this study. Thus, the \overline{rf} decreased from 145 ± 93 to 102 ± 50 when the data were normalized. This analysis can be used for other detectors which exhibit responses that are linear with changes in analyte concentration.

The empirical data also allows comparison of the resolving power of different detector arrays with the same number of detectors at known analyte concentrations. This allows comparison between the performance of different types of arrays without possible biases due to differing numbers of detectors in each type of system. Thus, the data obtained from the specific carbon black-polymer composite detector array having the median \overline{rf} of all possible combinations of twelve of the fourteen carbon black-polymer composite detectors studied are compared to the data produced by the available twelve element conducting polymer detector array. Similarly, arrays of carbon black-polymer composite, bulk organic conducting polymer, and tin oxide detectors are compared at a common array size of eight detector elements. In these comparisons, unnormalized data are used because the lack of linearity with varying analyte concentration of some of the detector modalities prevent meaningful normalization of the detector response data in those instances.

The carbon black-polymer composite detector array having the median \overline{rf} of all possible combinations of twelve of the fourteen carbon black-polymer composite detectors consist of polymers #2 and 4-14 (refer to the table in Figure 6). This particular array displayed $\overline{rf} = 117 \pm 79$ for the unnormalized response patterns to the set of test analytes at the fixed concentrations. For comparison, the available twelve element bulk organic conducting polymer array produce $\overline{rf} = 27 \pm 23$ for this same task.

The three detector arrays are also compared in resolving power using eight element arrays. The \overline{rf} value using unnormalized data for the eight tin oxide detectors

was 27 ± 23 , while that for the median-performing eight element carbon black-polymer composite array (detectors #3, 6-10, 12 and 13 from the table in Fig. 6) is 62 ± 32 and that of the median-performing eight element bulk conducting organic polymer array (detectors #1-4, 6, 7, 10 and 11 as labeled by the manufacturer Neotronics, Inc.) is 13 ± 10 . The trends in \overline{rf} values observed for twelve element arrays is thus also present at smaller array sizes, for the set of tasks and sets of detectors evaluated.

Rather than the average resolving power \overline{rf} , the resolution between the two poorest-resolved solvent pairs may be used to measure the performance of the arrays. To obtain this type of comparison, the array having the median rf_{worst} of all possible combinations of twelve of the fourteen carbon black-polymer composite detectors was compared to the rf_{worst} value of the available twelve detector conducting polymer array. The selected twelve element carbon black-polymer array contained detectors #1-3, 5 and 7-14 and displayed $rf_{\text{worst}} = 20$ (for anisole vs. butylamine), whereas the twelve element conducting polymer array yielded $rf_{\text{worst}} = 1.4$ (for n-heptane vs. α, α, α -trifluorotoluene). Using a set of only eight detectors allowed a comparison between all three types of detector modalities. Using this metric, the median-performing eight element carbon black-polymer composite array (detectors #2, 4, 5 and 8-12) had an $rf_{\text{worst}} = 11$ (anisole vs. butylamine) and the eight element tin oxide array had an rf_{worst} of 3.8 (for butylamine vs. triethylamine), while the median-performing eight element bulk conducting organic polymer array (detectors #1, 2, 4-6, 9, 10 and 12) had an $rf_{\text{worst}} = 1.0$ (n-heptane vs. α, α, α -trifluorotoluene).

The data can also be used to determine whether the combination of different sensor types realize significant improvement in the resolving power of the test solvents. The results of the combined arrays in resolving the nineteen analytes on average, and in resolving the worst-resolved pair, are shown in the table of Fig. 10. All pairwise combinations of complete (fourteen element carbon black-polymer composite, twelve element conducting organic polymer, and eight element tin oxide) arrays of different sensor type, are investigated, along with the performance of the full combination of all three arrays of different sensor types. To analyze the data in a consistent fashion for different arrays and vapors, autoscaled response data are used to formulate the resolution factors. The autoscaled response of the j th sensor to the i th exposure, A_{ij} , is:

$$A_{i,j} = \frac{(\Delta R_{ij,\max} / R_b) - \mu_j}{\eta_j}$$

where μ_j and η_j represent the mean and standard deviation, respectively, of the maximum relative differential resistance response of the j th sensor to the entire group of analytes.

A slight improvement in \overline{rf} from 9.7 to 9.8 is observed when the twelve
 5 available conducting organic polymer sensors are added to the fourteen element carbon black-polymer composite sensor array. The rf_{worst} increases more significantly, from 0.91 to 1.8, when these sensors are added. All other combinations of the full arrays, including the addition of the eight tin oxide sensors to the twenty-six sensor carbon black-polymer composite and conducting organic polymer sensors, underperform the twenty-six sensor
 10 array's performance for this set of test solvents using autoscaled sensor response data. An explanation of the observed performance improvement by adding the bulk organic conducting polymer sensors to the carbon black-polymer composite sensors is that the latter set is weakest at resolving some polar analytes, while the more polar bulk organic conducting polymer sensors are strong in resolving polar analytes. The combination of
 15 the two sensor types, in this case, improves the balance in chemical properties of the polymers in the sensor array and thus slightly improves the ability of the array to resolve this particular set of analytes.

The data can also be used to evaluate array performance as a function of the number of detectors in the array to determine how many detectors are required for
 20 various tasks. Both \overline{rf} and rf_{worst} are evaluated for all possible combinations of one through fourteen detectors in the carbon black-polymer composite detector array, one through eight detectors in the tin oxide array, and one through twelve detectors in the bulk organic conducting polymer array. One goal would be to construct the array that has the best average resolution between any pair of vapors that it might encounter. This criterion
 25 is closely related to the goal of constructing the array with the most chemical diversity, so that it best separates any set of vapors that is likely to be in the environment. To assess the diversity of the detector elements in a given detector modality, for all possible combinations of every n -element array of each detector type, the average pairwise resolution factor, $\overline{rf}(n)$, is computed for all 171 different solvent pairs of the nineteen
 30 solvents. The value of $\overline{rf}(n)$ for each allowed combination is tabulated, and the average value of $\overline{rf}(n)$ for all possible combinations having the specified number of detectors is plotted vs. the number of detectors in the array. The computed quantity is thus a

statistically-based measure of the average pairwise resolution of the solvents in the nineteen vapor test set that would be obtained from the average, unbiased and unsorted data arising from an n -element detector array of a given detector modality.

As displayed in Fig. 7A, these data show that, for all three detector arrays, the average resolving power of the array increased as the numbers of detectors in the array increased. Thus, for situations in which the sensing task is not known in advance, such as in quality control applications or in complex environments with varying backgrounds, the average performance of these tested detector arrays improved as the number of different detectors increased. For both raw response data and normalized data for the carbon black-polymer composite detectors, whose signal linearity vs. analyte concentration behavior permitted such an analysis to be performed in this test system, the same trend was observed for this particular array with this particular set of test analytes (as shown in Fig. 7B).

At any array size and for any of the detector arrays investigated, subsets of the full array that can outperform the mean $\overline{rf}(n)$ performance of all possible combinations of elements having that specific array size can always be identified. Analysis provides an answer to the question, given that the task to best resolve on average all nineteen specific test vapors, whether the best performing array contains the full collection of available detectors or instead only contains subsets of each detector modality. The quantity that is evaluated is the average pairwise resolution factor, determined using the Fisher linear discriminant method, of the best-performing set of detectors in an array, $\overline{rf}_{\max}(n)$, at each number of detectors, n . Fig. 7C depicts the $\overline{rf}_{\max}(n)$ data for each of the three detector modalities using raw response data. For each of the three detector modalities, a larger number of detectors provides increased resolving power according to the $\overline{rf}_{\max}(n)$ criterion.

In the case of the carbon black-polymer composite detectors, it is also possible to evaluate the concentration-independent performance using normalized data. As displayed in Fig. 7D, the optimum set of detectors obtained under these criteria once again is the full fourteen-detector array.

Another goal might be the selection of the subset of detectors that maximized the smallest pairwise solvent resolution factor for the analytes in the test set. The data in Fig. 8A describe the ability of each array type to resolve the worst-resolved analyte pair as a function of the size of the detector array. For each of the three detector

types, the resolving power increases rapidly and then appears to plateau at large numbers of detectors. However, in none of the three cases, using raw response data to analytes at fixed concentrations, does the addition of an extra detector diminish the resolving power of the array.

5 The array of eight tin oxide detectors maximizes the rf_{worst} at 3.8 (for butylamine vs. triethylamine). This rf_{worst} corresponds to approximately a 99% confidence in correctly distinguishing between the two analytes in a single presentation to the array. The array of twelve bulk conducting organic polymer detectors maximizes the rf_{worst} at 1.4 (for n-heptane vs. α,α,α -trifluorotoluene). This corresponds to
10 approximately an 84% confidence in correctly distinguishing between the two analytes on a single presentation to the array. The array of fourteen carbon black-polymer detectors maximizes rf_{worst} at 23. For this best set of fourteen detectors, the worst resolved analyte pair is anisole and butyl amine. The rf_{worst} of 23 essentially corresponds to a 100% confidence in distinguishing between the two analytes in a single presentation to
15 the detector array.

 Fig. 8B depicts rf_{worst} values for normalized data for the carbon black-polymer composite detectors. Again, the best resolution is maximized by the full fourteen detector array, although the data appears to plateau for sets containing more than eleven detectors. The full fourteen-detector set maximizes rf_{worst} at 21 (for butylamine
20 vs. n-pentane), which essentially corresponds to a 100% resolution probability. This full set of fourteen carbon black-polymer composite detectors contains polymers which vary widely in chemical properties, from the polar poly(N-vinylpyrrolidone) to the comparatively non-polar poly(ethylene-co-vinyl acetate), and includes halogenated and aromatic functional groups such as those in poly(vinylidene chloride-co-acrylonitrile) and
25 poly(9-vinylcarbazole). The diversity in the detector component properties, which matches well with the diversity in the analytes, leads to the ability to resolve all the nineteen analytes in the test set quite effectively.

 The present invention also allows analysis in the situation where the task is known and can be specified precisely in advance of array construction. Members of the
30 detector array can, in principle, be pre-selected to provide the optimum performance for the task of concern. The issue to be addressed is whether the optimal resolution of a specific solvent pair, at a fixed concentration, is obtained through use of the data produced by the full array, or whether it is advantageous, for that particular pair of

analytes, to use only the data produced by a subset of the full detector array. Ideally, all detectors would be orthogonal and one detector would probe exactly along the direction that is associated with maximizing the chemical differences between the two specific analytes. This single detector, and duplicates of it, should form the best "array" for resolving the two test analytes, under controlled conditions, at fixed concentrations. However, since in reality the vapor detectors are not generally mutually orthogonal, nor do they probe exactly the chemical differences between a specific analyte pair, it is probable that a few specific, partially correlated, detectors will combine to resolve a specific analyte pair better than any single detector in the array.

As a representative example, Fig. 9 depicts the results for resolving an arbitrarily chosen solvent pair, ethyl acetate from tetrahydrofuran, as a function of the size of the detector array, using only raw response data. All possible combinations of n detectors in the array are investigated for the fourteen detector carbon black-polymer composite array, the eight detector tin oxide array, and the twelve detector bulk conducting organic polymer array. Only elements from a given detector array are allowed to form the desired detector array (i.e., carbon black composites, tin oxide detectors, and bulk conducting organic polymers are restricted to be in mutually separate arrays for this analysis). For each specific number of detectors in the array, only the best resolution factors that resulted from the optimal array for the task of separating ethyl acetate from tetrahydrofuran, $rf_{best}(n)$, are plotted.

In each of the three cases, the ability of the arrays to resolve ethyl acetate from tetrahydrofuran increases as additional detectors are included. From Fig. 9, it is apparent that beyond approximately the first six carbon black-polymer detectors (#8-13 from the table in Fig. 6), additional detectors provide minimal increases in the resolution of ethyl acetate from tetrahydrofuran. Five of these six carbon black-polymer composite detectors contain polymers with one or both of the ether and ketone functionalities present in the ethyl acetate and tetrahydrofuran analytes, which explains chemically why the detectors perform well in this specific task. The other detectors in the array produce response data along vectors that do not lie as close to the vector that best separated these two solvents in detector space.

A similar metric is evaluated for all other solvent pairs as well. For each array class, using raw response data, the full array maximizes the resolution of each analyte pair. The best resolution factor that is obtained between any vapor pair, using any

set of detectors of the same class, is obtained for distinguishing n-pentane from toluene using the carbon black-polymer composite detector array which has $rf = 556$ using unnormalized data for this task. Similarly, using normalized data from the carbon black-polymer composite detector array, the best resolved analyte pair is dichloromethane from α, α, α -trifluorotoluene with $rf = 256$. The largest resolution factors obtained between any
5 analyte pair, using raw response data, for the conducting polymer or tin oxide detectors are 77 (for ethanol vs. isopropanol) and 148 (for chloroform vs. toluene), respectively, again for unnormalized data.

One of the findings to which the present invention leads is that array
10 performance increases, in general, as the number of sensors increases. The larger number of sensors is beneficial for resolving, on average, a generalized set of test vapors which might not all be known in advance of the array design. A larger number of sensors should increase the probability that the dimensionality of odor space is fully spanned by the array. Stated differently, the array should be able to probe many chemical differences
15 between analytes. On the other hand, with the analytes are fixed, a smaller number of selected sensors could optically resolve between solvent pairs. An optimal number of sensors can be determined for a given set of analytes.

The present invention is also capable of quantitatively pointing to the superiority of certain types of arrays for a given set of analytes. As between the tested
20 arrays of carbon black-polymer composite detectors, the tin oxide detectors, and conducting polymer detectors, it is clear that the carbon black-polymer composite detectors outperform the other sensor arrays. However, combining arrays of different detector of different modalities should lead to synergistic performance for the resolution of many analytes. For example, bulk organic conducting polymer detectors might be
25 placed in an array of surface acoustic wave (SAW) devices, tin oxide detectors, dye-impregnated polymer films on fiber optic detectors, quartz crystal microbalances (QCMs), or carbon black-polymer composite chemiresistors, for example. Besides the resolution performance, choosing between these various types of sensor arrays might require consideration of more practical matter, such as the signal-to-noise ratio of a given
30 type of a signal transduction mechanism, the ease of manufacturing the arrays and their costs, and the integration with processing electronics.

Hence the use of a numerical measurement of the resolution factor of a detector array according to the present invention can enhance the quantitative

understanding of the size of the sensor array, whether adding new detectors should be added, or whether certain chemical properties are underrepresented in a specific array, for example. The present invention helps the evaluation of other factors that are important in the design of arrays of vapor detectors.

5 It should be noted that the Fisher linear discriminant technique described herein is one analytical approach for resolving between analytes. Another related approach is principal component analysis in which the data are transformed such that they project along mutually orthogonal vectors in n-dimensional space so that the first principal component is the vector that captures the most variance in the data set and
10 higher principal components capture progressively less variance in the data. This statistical approach preserves the inherent weighing of the original data set in terms of molecular properties being probed by the sensors, but allows for improved visualization of the differences between the various patterns as they cluster in principal component space. In either principal component space or original, i.e., sensor, coordinate space
15 (used in the Fisher linear discriminant technique), a statistical measure of the differences between two clusters of patterns can be obtained by determining the distance between the centroids of the data arising from repeated exposures to a given vapor and dividing this distance by the sum of the standard deviations of the two different pattern clusters projected along the vector that connects these centroid points. This metric is independent
20 of the coordinate system chosen to display the data and is instead an inherent statistical property of the data set.

From standard vector analysis, the mean response vector, \mathbf{x}_a , of an n-sensor array to analyte a is given as the n-dimensional vector containing the mean autoscaled response of each sensors, A_{aj} , to the ath analyte as components such that

25
$$\mathbf{x}_a = (A_{a1}, A_{a2}, \dots, A_{an})$$

The average separation, $|\mathbf{d}|$, between the two analytes, a and b, in the Euclidean sensor response space is then equal to the magnitude of the difference between \mathbf{x}_a and \mathbf{x}_b . The noise of the sensor responses is also important in quantifying the resolving power of the sensor array. Thus the standard deviations, $\sigma_{a,d}$ and $\sigma_{b,d}$, obtained from all the individual
30 array responses to each of a and b along the vector \mathbf{d} , are used to describe the average separation and ultimately to define the pairwise resolution factor as

$$rf = \frac{|\bar{d}|}{\sqrt{\sigma_{a,\bar{d}}^2 + \sigma_{b,\bar{d}}^2}}$$

Further details of the present invention are described in the previously referenced U.S. Provisional Application No. 60/086,876, filed May 27, 1998 and U.S.

- 5 Provisional Application No. 60/088,804, filed June 9, 1998, which applications are hereby incorporated by reference for all purposes.

While the above is a complete description of the preferred embodiments of the invention, various alternatives, modifications and equivalents may be used. It should be evident that the present invention is equally applicable by making appropriate
10 modifications to the embodiments described above. Therefore, the above description should not be taken as limiting the scope of the invention which is defined by the appended claims.

WHAT IS CLAIMED IS:

- 1 1. A method of resolving different odors, said method comprising
2 disposing a plurality of d sensors in an array, each sensor having different
3 responses to said different odors;
4 exposing said d sensors to first and second odors;
5 generating first and second sets of data from each of said sensors, each set
6 corresponding to said first or second odor, each data capable of being represented by a
7 vector in a d-dimensional space;
8 determining an axis in said d-dimensional space, said axis having the property that
9 projections of said data points onto said axis in said d-dimensional space have optimal
10 separation on said axis; and
11 resolving said first odor from said second odor by said separation.
- 1 2. The method of claim 1 wherein d is greater than four.
- 1 3. The method of claim 2 wherein d is at least eight.
- 1 4. The method of claim 1 further comprising
2 quantifying said resolving step by a separation amount
- 1 5. The method of claim 1 wherein in said disposing step, each sensor
2 has an electrical response to said different odors.
- 1 6. The method of claim 1 further comprising
2 disposing said sensors on a single substrate.
- 1 7. The method of claim 6 wherein said substrate comprises
2 semiconductor material.
- 1 8. The method of claim 6 wherein said sensors comprise carbon
2 black-polymer composite sensors.
- 1 9. The method of claim 6 wherein said sensors comprise tin oxide
2 sensors.
- 1 10. The method of claim 6 wherein said sensors comprise bulk organic
2 conducting polymer sensors.

- 1 11. The method of claim 8 wherein said sensors sense odor by a
2 change in resistance responsive to exposure to said odor.
- 1 12. The method of claim 1 wherein said sensor disposing step
2 comprises
3 disposing a plurality of different types of sensors in an array.
- 1 13. The method of claim 12 wherein said disposing step further
2 comprises selecting at least one type of sensor from the group comprising surface
3 acoustic wave devices, tin oxide detectors, conducting organic polymers, dye-
4 impregnated polymer films on fiber optic detectors, polymer-coated micromirrors, quartz
5 crystal microbalances, electrochemical gas detectors, chemically sensitive field-effect
6 transistors, carbon black-polymer composite chemiresistors, micro-electro-mechanical
7 system devices, and micro-opto-electro-mechanical system devices.
- 1 14. The method of claim 13 wherein at least one type of sensor
2 comprises carbon black-polymer composite sensors.
- 1 15. The method of claim 13 wherein at least one type of sensor
2 comprises tin oxide sensors.
- 1 16. The method of claim 15 wherein at least one type of sensor
2 comprises bulk organic conducting polymer sensors.
- 1 17. The method of claim 16 wherein a second type of sensor comprises
2 carbon black-polymer composite sensors.
- 1 18. A system for resolving different odors, said system comprising
2 an array having a plurality of d sensors, each sensor having different electrical
3 responses to said different odors;
4 an array interface receiving first and second sets of data from each of said sensors
5 upon exposure of said d sensors to first and second odors, each set corresponding to said
6 first or second odor; and
7 a computer unit storing said first and second sets of data, said computer unit
8 processing each datum as a vector in a d-dimensional space, determining an axis in said
9 d-dimensional space, said axis having the property that projections of said data points

10 onto said axis in said d-dimensional space have optimal separation on said axis and
11 determining said optimal separation to resolve said first odor from said second odor by
12 said separation.

1 19. The system of claim 18 wherein said computer unit has a unit upon
2 which said optimal separation is displayed.

1 20. The system of claim 18 wherein said computer unit has a unit upon
2 which said optimal separation is stored.

1 21. The system of claim 18 wherein each sensor has an electrical
2 response to said different odors.

1 22. The system of claim 18 wherein said sensors are disposed on a
2 single substrate.

1 23. The system of claim 22 wherein said sensors comprise carbon
2 black-polymer composite sensors.

1 24. The system of claim 22 wherein said sensors comprise tin oxide
2 sensors.

1 25. The system of claim 22 wherein said sensors comprise bulk organic
2 conducting polymer sensors.

1 26. The system of claim 18 wherein said d sensors comprise a plurality
2 of different types of sensors.

1 27. The system of claim 26 wherein at least one type of sensor is
2 selected from the group comprising surface acoustic wave devices, tin oxide detectors,
3 conducting organic polymers, dye-impregnated polymer films on fiber optic detectors,
4 polymer-coated micromirrors, quartz crystal microbalances, electrochemical gas
5 detectors, chemically sensitive field-effect transistors, carbon black-polymer composite
6 chemiresistors, micro-electro-mechanical system devices, and micro-opto-electro-
7 mechanical system devices.

1 28. The system of claim 26 wherein at least one type of sensor
2 comprises carbon black-polymer composite sensors.

1 29. The system of claim 26 wherein at least one type of sensor
2 comprises tin oxide sensors.

1 30. The system of claim 26 wherein at least one type of sensor
2 comprises bulk organic conducting polymer sensors.

1 31. The system of claim 30 wherein a second type of sensor comprises
2 carbon black-polymer composite sensors.

1 32. A method of resolving different odors, said method comprising
2 disposing a plurality of sensors for exposure to said odors, said sensors
3 including at least two different types of sensors;
4 exposing said plurality of sensors to first and second odors; and
5 generating first and second sets of data from each of said sensors, each set
6 corresponding to said first or second odor; and
7 resolving said first odor from said second odor from an analysis of said
8 data sets.

1 33. The method of claim 30 wherein at least one type of sensor is
2 selected from the group comprising surface acoustic wave devices, tin oxide detectors,
3 conducting organic polymers, dye-impregnated polymer films on fiber optic detectors,
4 polymer-coated micromirrors, quartz crystal microbalances, electrochemical gas
5 detectors, chemically sensitive field-effect transistors, carbon black-polymer composite
6 chemiresistors, micro-electro-mechanical system devices, and micro-opto-electro-
7 mechanical system devices.

1 34. The method of claim 32 wherein one type of sensor comprises
2 carbon black-polymer composite chemiresistors and a second type of sensor comprises
3 conducting organic polymers.

1 35. A method of resolving different odors, said method comprising
2 exposing an array of odor sensors to a plurality of odors, each odor sensor
3 having an electrical response to at least one of said odors ;
4 determining the maximum relative electrical response for each odor and
5 sensor;

6 determining the average maximum relative electrical response for the
 7 determined maximum relative electrical responses for each odor;
 8 determining the standard deviation for the average maximum relative
 9 electrical response for the determined maximum relative electrical responses for each
 10 odor; and
 11 computing the optimal resolution factor by the Fisher linear discriminant
 12 for each pair of odors.

1 36. The method of claim 35 wherein the optical resolution factor *rf*
 2 equals

3
$$\frac{d_w}{\sqrt{\sigma_{a,w}^2 + \sigma_{b,w}^2}}$$

4 where *w* is a projection vector, $\sigma_{a,w}$ is the standard deviation for the average maximum
 5 relative response for odor a along vector *w*, $\sigma_{b,w}$ is the standard deviation for the average
 6 maximum relative response for the odor b along vector *w*, and *d_w* is the distance between
 7 the average maximum responses for odors a and b along vector *w*.

1 37. The use of Fisher linear discriminant optimum resolution factors
 2 from a set of data points obtained from exposures of an array of odor sensors to a
 3 plurality of odors to determine an optimum number of odor detectors for the array to
 4 resolve the plurality of odors.

1 38. The use of Fisher linear discriminant optimum resolution factors
 2 from a set of data points obtained from exposures of an array of different odor sensors to
 3 a plurality of odors to determine a subset of the odor sensors to resolve selected odors to a
 4 desired degree of certainty.

1 39. The use of Fisher linear discriminant optimum resolution factors
 2 from sets of data points obtained from exposures of arrays of different odor sensors to a
 3 plurality of odors to determine which array best resolves the plurality of odors.

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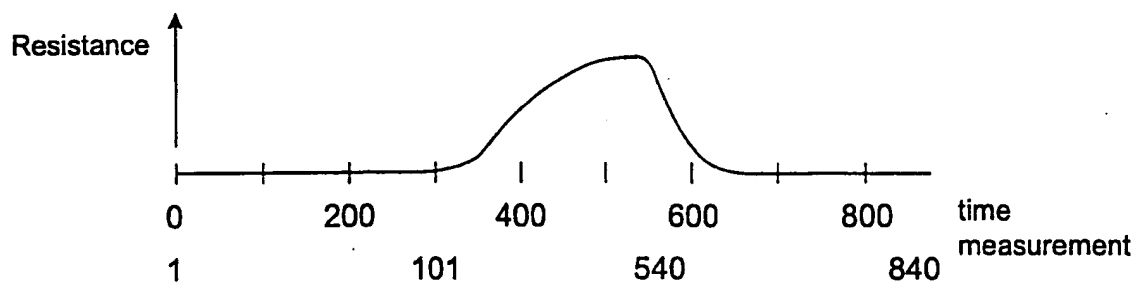
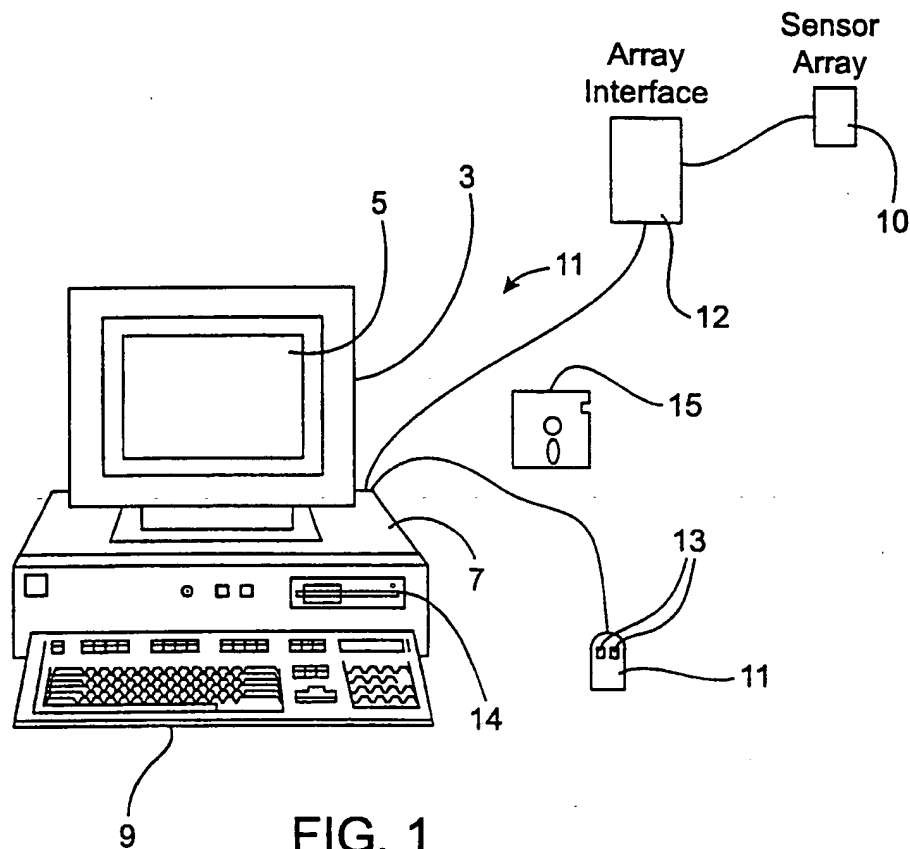


FIG. 2

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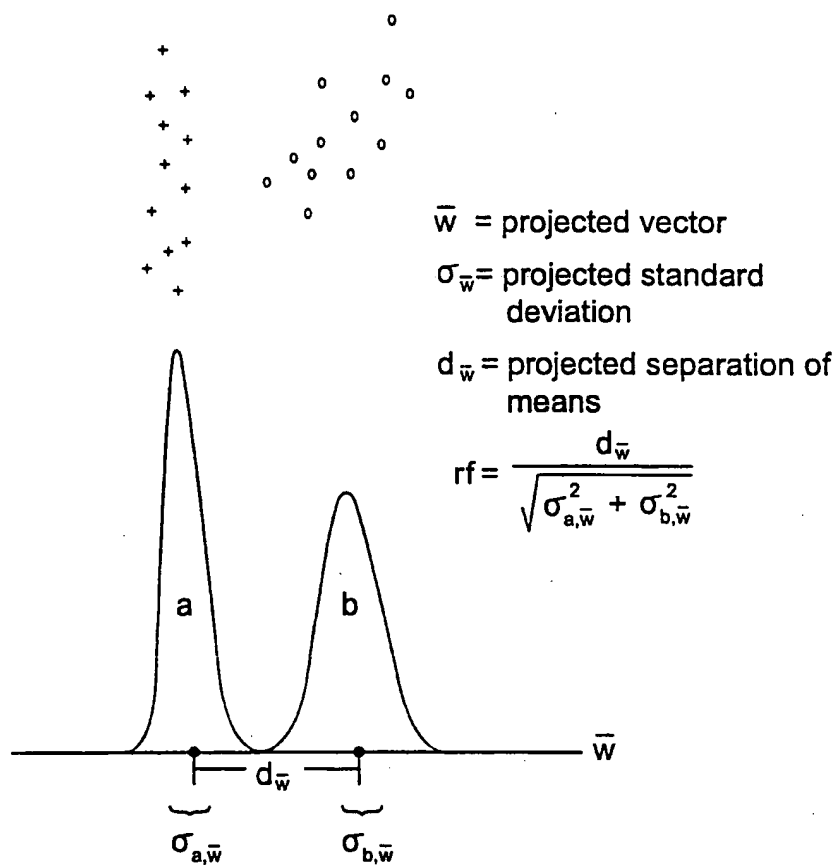


FIG. 3

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	acetone	acetonitrile	anisole	benzene	butylamine	chloroform	cyclohexane	dichloromethane	ethanol	ethyl acetate	isopropanol	methanol	n-heptane	n-pentane	tetrahydrofuran	toluene	triethylamine	α,α,α -trifluorotoluene
1,2-dimethoxyethane	121	194	71	49	58	95	287	130	135	164	184	230	156	354	130	179	73	119
acetone		51	102	98	44	106	88	42	58	93	107	41	95	172	42	227	138	198
acetonitrile			154	173	54	153	152	73	58	171	207	75	119	135	88	317	174	200
anisole				106	23	178	146	130	107	126	131	129	140	182	145	59	78	60
benzene					52	65	177	132	117	58	140	202	112	195	102	229	47	82
butylamine						55	63	53	48	36	82	86	76	54	48	41	57	65
chloroform							123	102	122	133	126	204	92	231	159	303	146	180
cyclohexane								109	103	180	108	151	202	152	153	528	132	312
dichloromethane									57	121	60	72	132	54	63	489	246	384
ethanol										119	102	57	77	77	51	279	125	137
ethyl acetate											88	201	123	181	59	253	122	151
isopropanol												135	161	119	73	313	218	206
methanol													107	58	87	401	308	260
n-heptane														177	161	380	156	90
n-pentane															217	556	171	343
tetrahydrofuran																359	93	178
toluene																	246	180
triethylamine																		168

14 Carbon Black- Polymer Composite Detectors

FIG. 4A

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	acetone	acetonitrile	anisole	benzene	butylamine	chloroform	cyclohexane	dichloromethane	ethanol	ethyl acetate	isopropanol	methanol	n-heptane	n-pentane	tetrahydrofuran	toluene	triethylamine	α,α,α -trifluorotoluene
1,2-dimethoxyethane	18	48	102	10	7.3	120	44	106	7.3	35	18	15	5.5	64	33	26	7.2	22
acetone		36	59	6.5	11	50	10	47	8.1	15	11	12	10	11	17	27	16	19
acetonitrile			57	7.3	29	103	13	94	49	14	26	21	21	26	23	15	33	49
anisole				23	45	53	37	53	67	54	57	48	43	56	41	54	66	51
benzene					13	9.7	7.0	12	7.8	8.6	8.3	6.9	4.9	5.2	5.1	11	14	9.7
butylamine						37	36	36	10	26	16	12	9.1	31	22	22	3.8	30
chloroform							37	20	24	34	47	44	27	38	16	149	38	7.9
cyclohexane								25	10	13	22	19	12	6.6	8.8	19	24	8.9
dichloromethane									25	32	48	35	27	25	20	116	36	4.9
ethanol										16	12	11	7.2	20	20	34	11	12
ethyl acetate											13	14	19	17	12	21	21	17
isopropanol												15	7.7	53	25	16	16	19
methanol													11	14	11	16	20	26
n-heptane														23	25	9.1	12	28
n-pentane															14	29	27	16
tetrahydrofuran																17	26	9.6
toluene																	26	48
triethylamine																		23

8 Tin Oxide Detectors

FIG. 4B

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	acetone	acetonitrile	anisole	benzene	butylamine	chloroform	cyclohexane	dichloromethane	ethanol	ethyl acetate	isopropanol	methanol	n-heptane	n-pentane	tetrahydrofuran	toluene	triethylamine	α,α,α -trifluorotoluene
1,2-dimethoxyethane	5.7	24	9.7	7.3	10	9.5	8.7	6.5	46	5.5	25	29	9.8	11	9.5	7.7	21	8.4
acetone		22	15	14	11	16	19	6.7	64	12	35	29	20	18	14	10	19	8.7
acetonitrile			36	35	23	45	32	23	54	27	36	28	41	39	45	31	43	34
anisole				3.1	12	4.4	7.1	12	26	6.9	17	42	8.5	9.5	14	3.5	9.9	5.0
benzene					13	2.7	2.8	4.3	18	5.0	7.9	31	3.3	4.4	8.8	3.5	9.8	2.0
butylamine						12	14	11	35	17	12	40	12	11	11	12	18	12
chloroform							6.1	6.8	37	6.2	8.4	35	7.3	12	10	3.4	10	5.1
cyclohexane								3.5	66	7.5	22	27	1.6	4.3	9.8	4.5	12	2.6
dichloromethane									57	4.0	15	31	4.0	5.7	4.8	5.2	14	3.5
ethanol										62	78	55	49	46	42	16	49	24
ethyl acetate											20	29	11	9.4	11	4.9	14	5.2
isopropanol												39	12	21	10	8.0	39	6.4
methanol													35	41	38	30	38	34
n-heptane														4.4	5.5	4.7	9.2	1.4
n-pentane															9.3	5.2	15	4.2
tetrahydrofuran																4.9	16	5.5
toluene																	13	4.4
triethylamine																		9.7

12 Bulk Conducting Organic Polymer Detectors

FIG. 4C

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	acetone	acetonitrile	anisole	benzene	butylamine	chloroform	cyclohexane	dichloromethane	ethanol	ethyl acetate	isopropanol	methanol	n-heptane	n-pentane	tetrahydrofuran	toluene	triethylamine	α,α,α -trifluorotoluene
1,2-dimethoxyethane	59	107	76	61	59	110	120	41	68	103	58	61	157	129	74	111	97	138
acetone		30	74	88	29	86	103	56	47	59	77	32	93	64	58	82	105	148
acetonitrile			154	176	70	145	167	81	63	107	208	40	133	99	91	225	171	182
anisole				40	23	134	91	117	90	106	88	126	128	100	119	48	102	58
benzene					54	92	100	88	115	84	144	168	113	99	136	31	51	43
butylamine						55	75	65	37	35	67	54	116	21	31	31	74	77
chloroform							74	89	97	139	95	166	154	89	146	118	142	156
cyclohexane								111	159	120	107	151	85	107	107	141	40	71
dichloromethane									63	48	59	70	110	45	42	229	197	256
ethanol										55	90	30	79	60	50	189	112	111
ethyl acetate											124	149	185	50	42	157	158	143
isopropanol												89	181	55	84	123	203	141
methanol													74	54	71	187	240	149
n-heptane														98	221	187	203	82
n-pentane															56	119	70	109
tetrahydrofuran																184	112	149
toluene																	82	27
triethylamine																		96

FIG. 5

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Sensor #	Polymer
1	poly(4-vinyl phenol)
2	poly(vinyl chloride-co-vinyl acetate), 10% vinyl acetate
3	poly(<i>N</i> -vinylpyrrolidone)
4	poly(vinyl acetate)
5	poly(methyl vinyl ether-co-maleic anhydride)
6	poly(carbonate bisphenol A)
7	poly(styrene)
8	poly(sulfone)
9	poly(methyl methacrylate)
10	poly(vinylidene chloride-co-acrylonitrile), 80% vinylidene chloride
11	poly(caprolactone)
12	poly(ethylene-co-vinyl acetate), 82% ethylene
13	poly(ethylene oxide)
14	poly(9-vinylcarbazole)

FIG. 6

		carbon black-polymer	conducting polymer	tin oxide
carbon black-polymer	average rf	9.7	9.8	9.0
	worst rf	0.91	1.8	0.88
conducting polymer	average rf		5.2	7.2
	worst rf		0.30	0.62
tin oxide	average rf			5.8
	worst rf			0.46
carbon black-polymer, conducting polymer, and tin oxide arrays			average rf:	9.4
			worst rf:	0.95

FIG. 10

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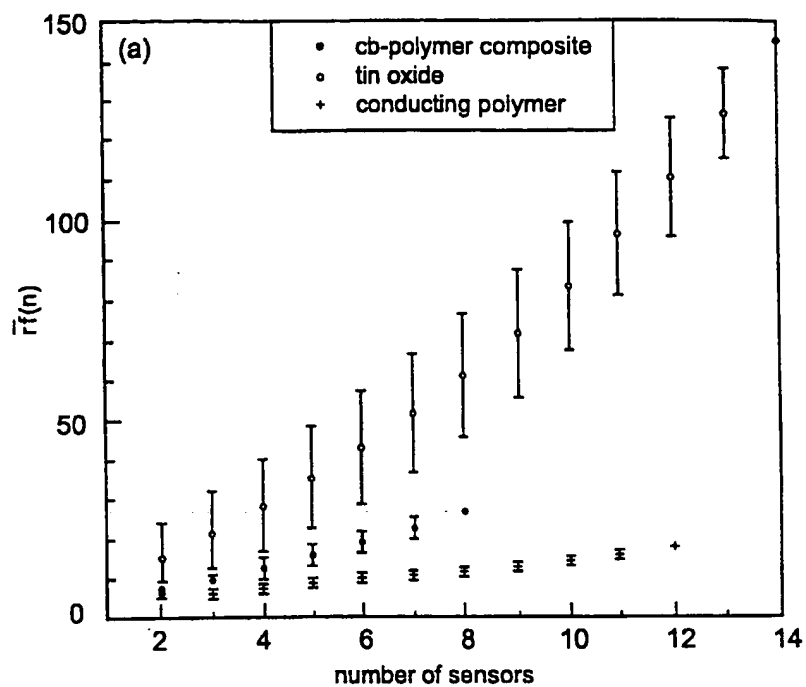


FIG. 7A

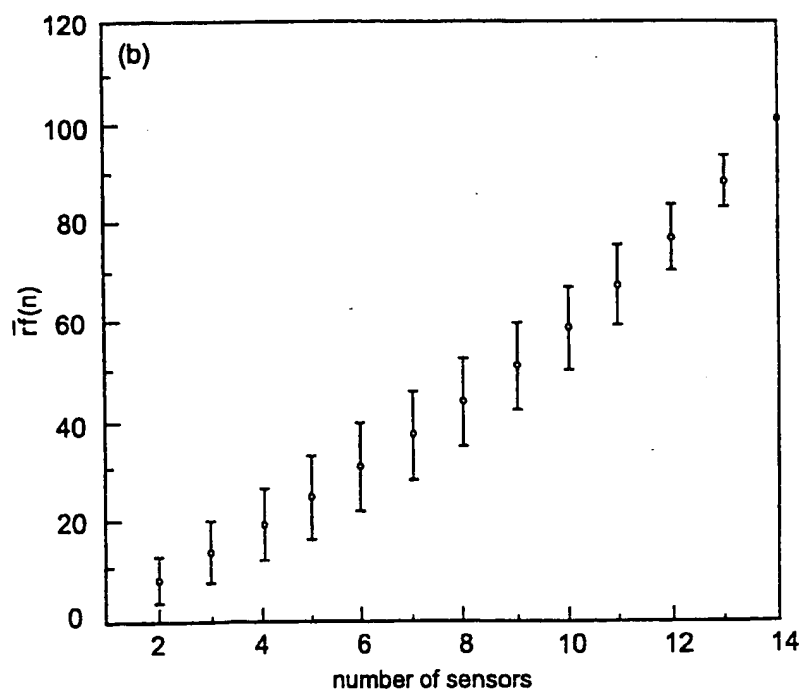


FIG. 7B

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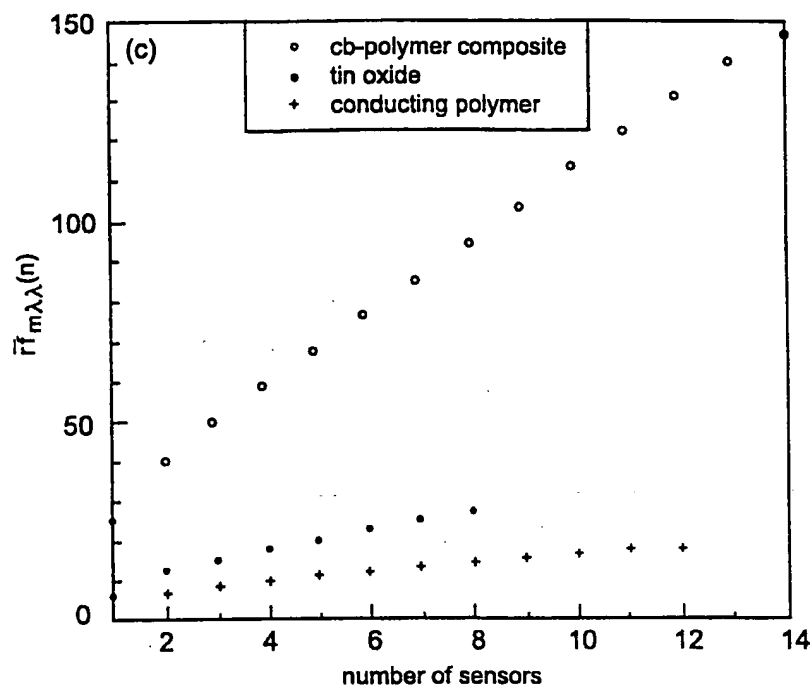


FIG. 7C

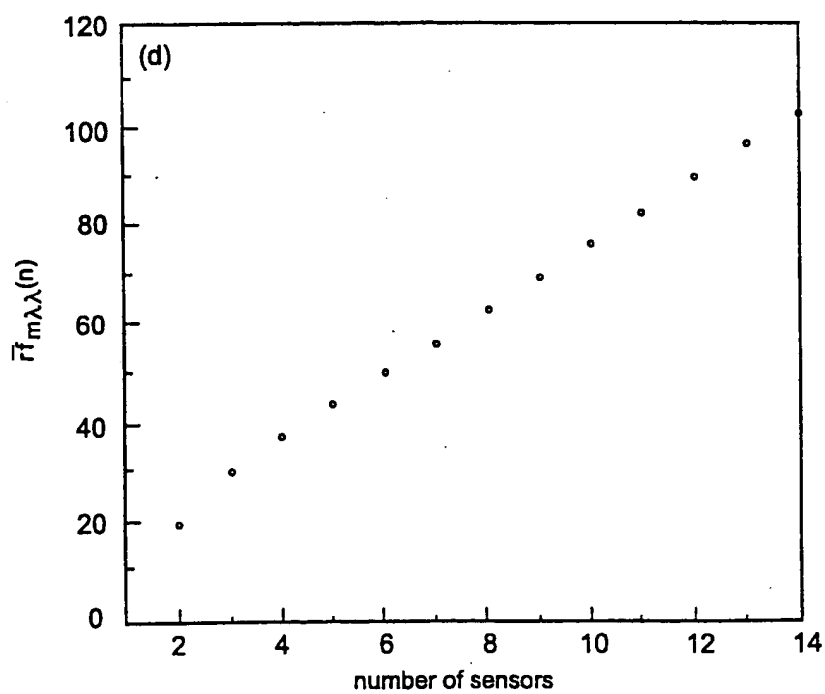


FIG. 7D

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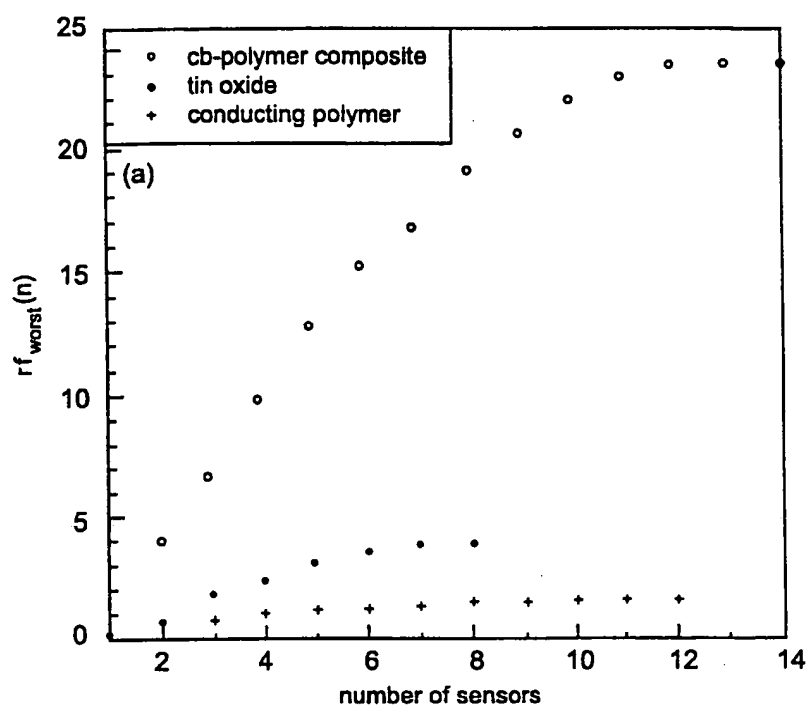


FIG. 8A

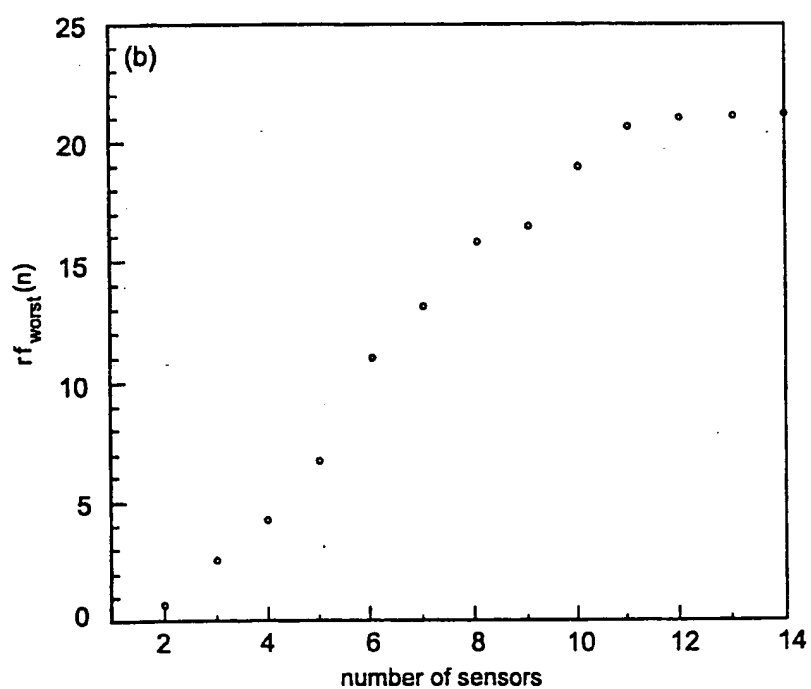


FIG. 8B

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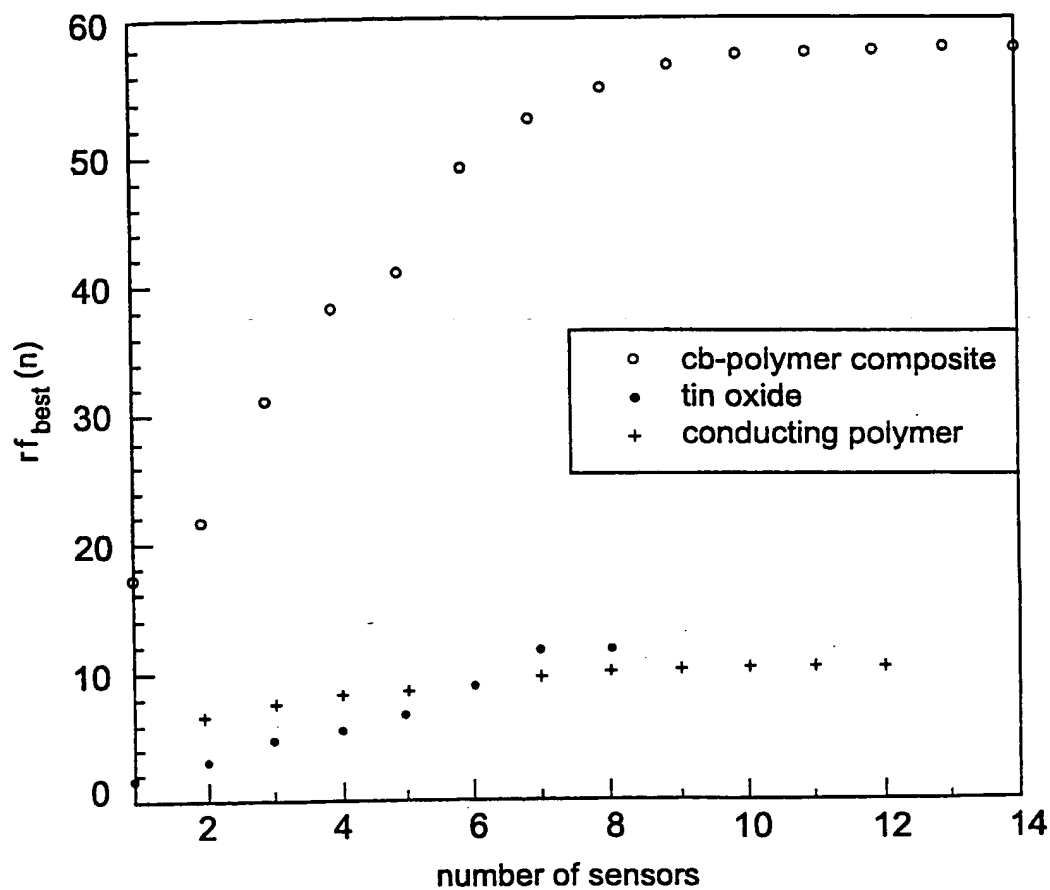


FIG. 9

INTERNATIONAL SEARCH REPORT

International application No.
PCT/US99/11636

A. CLASSIFICATION OF SUBJECT MATTER

IPC(6) :G01N 27/27

US CL :422/83, 88, 90, 98; 436/149, 151, 181; 73/23.34; 702/27, 30, 32

According to International Patent Classification (IPC) or to both national classification and IPC

B. FIELDS SEARCHED

Minimum documentation searched (classification system followed by classification symbols)

U.S. : 422/83, 88, 90, 98; 436/149, 151, 181; 73/23.34; 702/27, 30, 32

Documentation searched other than minimum documentation to the extent that such documents are included in the fields searched

Electronic data base consulted during the international search (name of data base and, where practicable, search terms used)

Please See Extra Sheet.

C. DOCUMENTS CONSIDERED TO BE RELEVANT

Category*	Citation of document, with indication, where appropriate, of the relevant passages	Relevant to claim No.
X	US 5,469,369 A (ROSE-PEHRSSON ET AL) 21 November 1995	1-34
--	(21.11.95), column 1, lines 26-30; column 3, lines 1-11; column 3,	-----
Y	line 61 - column 4, line 3; column 8, lines 35-67; column 13, lines	35-39
	41-58; and column 15, lines 9-11 and 26-39.	
Y	US 5,654,497 A (HOFFHEINS ET AL) 05 August 1997 (05.08.97),	1-39
	column 4, lines 20-39; column 4, line 66 - column 5, line 14;	
	column 5, lines 58-69; and column 8, lines 3-58.	



Further documents are listed in the continuation of Box C.



See patent family annex.

* Special categories of cited documents:	*T* later document published after the international filing date or priority date and not in conflict with the application but cited to understand the principle or theory underlying the invention
A document defining the general state of the art which is not considered to be of particular relevance	*X* document of particular relevance; the claimed invention cannot be considered novel or cannot be considered to involve an inventive step when the document is taken alone
B earlier document published on or after the international filing date	*Y* document of particular relevance; the claimed invention cannot be considered to involve an inventive step when the document is combined with one or more other such documents, such combination being obvious to a person skilled in the art
L document which may throw doubts on priority claim(s) or which is cited to establish the publication date of another citation or other special reason (as specified)	*G* document member of the same patent family
O document referring to an oral disclosure, use, exhibition or other means	
P document published prior to the international filing date but later than the priority date claimed	

Date of the actual completion of the international search

17 AUGUST 1999

Date of mailing of the international search report

10 SEP 1999

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INTERNATIONAL SEARCH REPORT

International application No.
PCT/US99/11636

C (Continuation). DOCUMENTS CONSIDERED TO BE RELEVANT

Category*	Citation of document, with indication, where appropriate, of the relevant passages	Relevant to claim No.
Y	Database CAPLUS on STN, American Chemical Society, (Columbus, OH, USA), Abstract No. 1996:445460, ODORICO, R. 'Neural 2.00-A program for neural net and statistical pattern recognition,' abstract, Comput. Phys. Commun., Vol. 96, No. 2&3, pages 314-330, issued 1996.	1-39

INTERNATIONAL SEARCH REPORT

International application No.
PCT/US99/11636

B. FIELDS SEARCHED

Electronic data bases consulted (Name of data base and where practicable terms used):

APS, CAPLUS

search terms: Fisher? line? discrimin?; Fisher? line?; line? discrim?; (odor? or gas### or analyt?) (2a) (detn or determin?
or detect? or measur? or sens###)